

Spyder (Python 3.9)

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D:\Master2022_2023\Sim2\Information System\ATIS Python code\Recommender Systems.py

```
1 # Applied Recommender Systems with Python
2 #Hello World of a Simple Recommender System
3
4 # Import Pandas
5 import pandas as pd
6
7 # Load Movies Metadata
8 metadata = pd.read_csv('movies_metadata.csv', low_memory=False)
9
10 # Print the first three rows
11 metadata.head(3)
12
13 # First step: Mean rating
14
15 # Calculate mean of vote average column
16 C = metadata['vote_average'].mean()
17 print('vote average: ', C)
18
19 # First step: Number of votes
20
21 # Calculate the minimum number of votes required to be in the chart, m
22 m = metadata['vote_count'].quantile(0.90)
23 print('minimum number of votes required: ', m)
24
25 # First step: DataFrame
26
27 # Filter out all qualified movies into a new DataFrame
28 q_movies = metadata.copy().loc[metadata['vote_count'] >= m]
29 q_movies.shape
30 metadata.shape
31
32 # First step: Feature score
33
34 # Function that computes the weighted rating of each movie
35 def weighted_rating(x, m=m, C=C):
36     v = x['vote_count']
37     R = x['vote_average']
38     # Calculation based on the IMDB formula
39     return (v/(v+m) * R) + (m/(m+v) * C)
40
41 # Define a new feature 'score' and calculate its value with 'weighted_rating'
42 q_movies['score'] = q_movies.apply(weighted_rating, axis=1)
43
44 # First step: Sort the DataFrame
45
46 #Sort movies based on score calculated above
47 q_movies = q_movies.sort_values('score', ascending=False)
48
49
```

Name	Type	Size	Value
C	float	1	5.618207215133889
m	float64	1	160.0
metadata	DataFrame	(45466, 24)	Column names: adult, belongs_to_collection, budget, genres, homepage, ...
q_movies	DataFrame	(4555, 25)	Column names: adult, belongs_to_collection, budget, genres, homepage, ...
tfidf	feature_extraction.text.TfidfVectorizer	1	TfidfVectorizer object of sklearn.feature_extraction.text module
tfidf_matrix	sparse_csr.csr_matrix	(45466, 75827)	csr_matrix object of scipy.sparse.csr module

Help Variable Explorer Plots Files

Console I/A X

```
In [5]: runfile('D:/Master2022_2023/Sim2/Information System/ATIS Python code/Recommender Systems.py', wdir='D:/Master2022_2023/Sim2/Information System/ATIS Python code')
vote average: 5.618207215133889
minimum number of votes required: 160.0
C:\Users\WJkSoft\Anaconda3\lib\site-packages\sklearn\utils\deprecation.py:87: FutureWarning: Function get_feature_names is deprecated; get_feature_names is deprecated in 1.0 and will be removed in 1.2. Please use
```

LSP Python: ready conda: base (Python 3.9.13) Line 85, Col 51 ASCII CRLF RW Mem 59%

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46 #Sort movies based on score calculated above
47 q_movies = q_movies.sort_values('score', ascending=False)
48
49
```

Index	adult	belongs to collection	budget	genres
0	False	{'id': 10194, 'name': 'Toy Story Collection...	300000000	[{'id': 16, 'name': 'Animation'}, {'id': 35, 'name': 'Comedy'}]
1	False	nan	650000000	[{'id': 12, 'name': 'Adventure'}, {'id': 14, 'name': 'Romance'}, {'id': 35, 'name': 'Comedy'}]
2	False	{'id': 119050, 'name': 'Grumpy Old Men Coll...	0	[{'id': 10749, 'name': 'Romance'}, {'id': 35, 'name': 'Comedy'}]
3	False	nan	160000000	[{'id': 35, 'name': 'Comedy'}, {'id': 18, 'name': 'Horror'}]
4	False	{'id': 96871, 'name': 'Father of the Bride ...	0	[{'id': 35, 'name': 'Comedy'}]
5	False	nan	600000000	[{'id': 28, 'name': 'Action'}, {'id': 80, 'name': 'Romance'}]
6	False	nan	580000000	[{'id': 35, 'name': 'Comedy'}, {'id': 10749, 'name': 'Romance'}]
7	False	nan	0	[{'id': 28, 'name': 'Action'}, {'id': 12, 'name': 'Horror'}]
8	False	nan	350000000	[{'id': 28, 'name': 'Action'}, {'id': 12, 'name': 'Horror'}]
9	False	{'id': 645, 'name': 'James Bond Collection...	580000000	[{'id': 12, 'name': 'Adventure'}, {'id': 28, 'name': 'Action'}, {'id': 35, 'name': 'Comedy'}]
10	False	nan	620000000	[{'id': 35, 'name': 'Comedy'}, {'id': 18, 'name': 'Horror'}]
11	False	nan	0	[{'id': 35, 'name': 'Comedy'}, {'id': 27, 'name': 'Horror'}]
12	False	{'id': 117693, 'name': 'Balto Collection', ...	0	[{'id': 10751, 'name': 'Family'}, {'id': 16, 'name': 'Animation'}]
13	False	nan	440000000	[{'id': 35, 'name': 'History'}, {'id': 18, 'name': 'Drama'}]

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LSP Python: ready conda: base (Python 3.9.13) Line 85, Col 51 ASCII CRLF RW Mem 61%

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untitled0.py x Recommender Systems.py x extrath.py x

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17 C = metadata['vote_average'].mean()
18 print('vote average: ', C)
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22 # Calculate the minimum number of votes
23 m = metadata['vote_count'].quantile(0.95)
24 print('minimum number of votes: ', m)
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26 # First step: DataFrame
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29 q_movies = metadata.copy().loc[metadata['vote_count'] >= m]
30 q_movies.shape
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38     R = x['vote_average']
39     # Calculation based on the formula
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44
45 # First step: Sort the DataFrame
46
47 # Sort movies based on score calculated
48 q_movies = q_movies.sort_values('score', ascending=False)
49
```

Name	Type	Size	Value
C	float	1	5.618287215133889
m	float64	1	168.0
metadata	DataFrame	(45466, 24)	Column names: adult, belongs_to_collection, budget, genres, homepage, ...
q_movies	DataFrame	(4595, 25)	Column names: adult, belongs_to_collection, budget, genres, homepage, ...
tfidf	feature_extraction.text.TfidfVectorizer	1	TfidfVectorizer object of sklearn.feature_extraction.text module
tfidf_matrix	sparse.csr.csr_matrix	(45466, 75827)	csr_matrix object of scipy.sparse.csr module

q_movies - DataFrame

Index	adult	belongs to collection	budget	genres	homepage
314	False	nan	25000000	[{'id': 18, 'name': 'Drama'}, {'id': 80, 'name': 'Crime'}]	nan
834	False	{'id': 230, 'name': 'The Godfather Collecti..	60000000	[{'id': 18, 'name': 'Drama'}, {'id': 80, 'name': 'Crime'}]	http://
10309	False	nan	13200000	[{'id': 35, 'name': 'Comedy'}, {'id': 18, 'name': 'Drama'}]	http://
12481	False	{'id': 263, 'name': 'The Dark Knight Collec..	185000000	[{'id': 18, 'name': 'Drama'}, {'id': 28, 'name': 'Action'}]	http://
2843	False	nan	63000000	[{'id': 18, 'name': 'Drama'}]	http://
292	False	nan	80000000	[{'id': 53, 'name': 'Thriller'}, {'id': 80, 'name': 'Crime'}]	http://
522	False	nan	22000000	[{'id': 18, 'name': 'Drama'}, {'id': 36, 'name': 'Animation'}]	http://
23673	False	nan	33000000	[{'id': 18, 'name': 'Drama'}]	http://
5481	False	nan	15000000	[{'id': 14, 'name': 'Fantasy'}, {'id': 12, 'name': 'Adventure'}]	http://
2211	False	nan	20000000	[{'id': 35, 'name': 'Comedy'}, {'id': 18, 'name': 'Drama'}]	http://
1178	False	{'id': 230, 'name': 'The Godfather Collecti..	13000000	[{'id': 18, 'name': 'Drama'}, {'id': 80, 'name': 'Crime'}]	http://
1152	False	nan	30000000	[{'id': 18, 'name': 'Drama'}]	http://
351	False	nan	55000000	[{'id': 35, 'name': 'Comedy'}, {'id': 18, 'name': 'Drama'}]	http://
1154	False	{'id': 10, 'name': 'Star Wars Collection', ...	180000000	[{'id': 12, 'name': 'Adventure'}, {'id': 28, 'name': 'Action'}]	http://

Warning: Function moved in 1.2. Please use

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